

Cybermycelium: Domain Driven Distributed Reference Architecture For Big Data Systems

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Abstract

The ubiquity of digital devices, the infrastructure of today, and the ever increasing proliferation of digital products have dawned a new era, the era of big data. This era began when the volume, variety and velocity of data overwhelmed traditional systems that used to analyze and store that data. This precipitated a new class of software systems, namely big data systems. Whereas big data systems provide competitive advantage to businesses, many failed to harness the power of it. It has been estimated that only 20% of companies have successfully implemented a big data project. This is due to various challenges of adopting big data such as organizational culture, rapid technology change, system development, data architecture and data engineering. This paper aims to facilitate big data system development, architecture and data engineering by introducing a domain driven decentralized big data reference architecture. This artefact is developed following the guidelines of empirically grounded reference architectures, and has been evaluated in a real world scenario. At the end, a prototype of the artefact has been instantiated and utilized to solve a real problem in practice. Our results displayed a good degree of applicability and some thought-provoking architectural tradeoffs and challenges.

Keywords: Reference architecture, Architecture, Big data reference architecture, Big data architecture, Big data systems, Big data software engineering,

1. Introduction

Since the dawn of internet and world wide web, humanity has witnessed a degree of connection beyond reckoning. The proliferation of digital devices pervaded with powerful software applications have created cyber communities that constantly mutate [1]; [2]. In a world where we have network infrastructures that can support up to 250Mbps of data transmission, and smart phones and IOT devices that can have processing power of up to 3 Ghz, data becomes ubiquitous, the quantum that lays the foundation of the nexus [3].

According to InternetLiveStates.com [4], only in one second, there are 9,878 tweets sent, 1,138 instagram photos uploaded, 3,117,720 emails sent, 99,738 Google searches made, and 94,144 Youtube videos viewed. That is, if it has taken 5 second to read the preceding paragraph, during that time, 15,588,600 emails are sent.

Driven by the ambition to harness the power of this deluge of data, the term 'Big Data' (BD) was coined [5]. BD initially emerged to address the challenges associated with various characteristics of data such as velocity, variety, volume and variability [2]. BD is the practice of extracting patterns, theories, and predictions from a large set of structured, semi-structured, and unstructured data for the purposes of business competitive advantage [6]; [7]. BD is a game-changing innovation, heralding the dawn of a new data-oriented industry.

Nonetheless, BD is not a magical wand that can enchant any business process. While a lot of opportunities exist in BD, subsuming an emergent and rather high-impacting technology like BD to current state of affairs in organizations, is a daunting task. According to recent survey from Databricks, only 13% of the organizations excel at delivering on their data strategy [8]. Another survey by NewVantage Partners indicated that only 24% organization have successfully gone data-driven [9]. This survey also states that only 30% of organizations have a well established strategy for their BD endeavour. In addition, surveys from McKinsey & Company ([10]) and Gartner ([11]) further support these numbers, which illuminates on the scarcity of successful BD implementations in

the industry.

There are various challenge for adoption of big data such as 'data architecture', 'rapid technology change', 'lack of sufficiently skilled data engineers', and 'organization's cultural challenges of becoming data-driven' [2];[12]. Among
35 these challenges, 'data architecture' is highlighted. A successful big data system is built upon a solid data architecture, serving as a blue print that affects data lifecycle management, guides data integration, control data assets and handle change.

Nevertheless, majority of BD systems are developed on-premise as ad-hoc
40 complicated solutions that do not adhere to the practices of software engineering and software architecture [13]; [14]. As the ecosystem grows and new technologies and data processing techniques are introduced, the software architect will have a harder time to come up with a solution that address the problem requirements.

This can potentially create grounds for an immature architecture that results in solutions that are hard to scale, hard to maintain, and raise high-entry
45 blockades [3]. Since the approach of ad-hoc design to BD system development is not desirable and may leave many architects and data engineers in the dark, novel data architectures that are designed specifically for BD are required. To contribute to this goal, we explore the notion of reference architectures (RAs)
50 and present a distributed domain-driven software RA for BD systems. This RA is called Cybermycelium. The contribution of this work is threefold: 1) explicating design theories underlying current big data systems, 2) explicating design theories that generates the artifacts constructs, and 3) the artifact.

The contents of this paper is presented in the following order: 1) in *back-*
55 *ground* section we provide a brief discussion on what's known about the topic under investigation, the problems associated with it, and the rationale for our study, 2) in *research methodology* section we discuss the integral elements of our research methodology with corresponding justifications, 3) in *requirement*
60 *specification* section, we discuss the requirements that has led to creation of the artifact, 4) in *why Cybermycelium* section, we expand on the problem stated in

the background section and provide with theories that explains the challenges of maintaining current BD systems, 5) in *Cybermycelium: A Domain Driven Distributed Reference Architecture for Big Data Systems* section we explicate the theories that underpin the artifact development and the artifact itself, 6) in *Evaluation* section, we provide an evaluation of the artifact, 7) in *discussion* section we elaborate on other BD RAs, and the distinct qualities and limitations of our artifact, 8) in *conclusion* we conclude the study summing up the findings.

2. Background

In this section, we provide a brief discussion on what is known about BD architectures, articulate the research gap and problems that need addressing, and present with the objective of this research.

2.1. Big Data Architectures: State of the art

The available body of knowledge and the knowledge from practice highlight 3 generations of BD architectures;

1. **Enterprise Data Warehouse:** this is perhaps one of the oldest approaches to business intelligence and data crunching and has existed even before the term BD was coined [15]. Usually developed as proprietary software, this data architecture pivot on enterprise data warehouse, ETL jobs, and a data visualization software such as Microsoft Power BI. As the data sources and consumers grow, this architecture suffers from hard to maintain ETL jobs, and visualizations that can be created and understood by a certain group of stakeholders, hindering data's positive impact on business. This also means, new transformations will take longer to be added to the workload, the system is monolithic and hard to scale, and only a few group of hyper-specialized individuals are able to operate the system. Moreover, data warehouses have been designed with different assumptions that cannot effectively handle the characteristics of big data.

2. **Data Lake:** to address the challenges occurred in the first generation of data architectures, a new BD ecosystem emerged. This new ecosystem revolved around a data lake, in a way that there isn't as much transformations on the data, but rather everything is dumped into the data lake and retrieved when necessary. Although data lake architecture have reached a higher level of success in comparison to the first generation of the data architectures, they are still far from optimal. As data consumer and data providers grow, data engineers will be immensely challenged to avoid creation of data swamp [16], and because there is usually no concept of data owner, the whole stack is usually operated by a group of hyper-specialized data engineers, creating silos, and barriers for gradual adoption. This also means various teams will often go into data engineers backlog and will not be in control of how and when they can consume the data they desire. Furthermore, data engineers are usually oblivious of the semantics and value of the data they are processing; they simply do not know how is that data useful or which domain it belongs to. This will overtime decrease the quality of data processing, results in haphazard data management, and make maintenance and data engineering a complicated task.
3. **Cloud Based Solutions:** Given the cost and complexity of running a data lake on-premise alongside the whole data engineering pipeline, and the substantial talent gap currently faced in the market [6], the third generation of BD architectures tend to revolve around as-a-service or on-demand cloud-based solutions. This generation of architecture tends to be leaning towards stream-processing with architectures such as Kappa or Lambda [17], or frameworks that unify batch and stream processing such as Apache Beam [18] or Databricks [19]. This is usually accompanied by cloud storage such as Amazon S3, and streaming technologies such as Amazon Kinesis. Whereas this generation tends to solve various issues regarding the complexity and cost of data handling and digestion, it still suffers from the same fundamental architectural challenges. It does not

120 have clear data domains, a group of siloed hyper-specialized data engineers
are running them, and data storage through a monolithic data pipelines
soon becomes a choke-point.

To discuss the integral facets that embroil these architectures, one must
look at the characteristics of these architectures and the ways in which they
125 achieve their ends, with quality attributes surrounding it. Except for one case
[3], all the architectures and RAs found as the result of this study, were designed
underlying monolithic data pipeline architecture with four major components
being data consumer, data processing, data infrastructure and data providers.

The process of turning data into actionable insights in these architectures
130 usually follow a similar lifecycle;

1. **Data Ingestion:** system beings to ingest data from all corners of the
enterprise, including both transactional, operational and external data.
For instance, in a practice management software for veterinaries, data
platform can ingest and persist transactional data such as 'user interaction
135 with therapeutics', 'number of animals diagnosed', or 'number of invoices
created' and 'medicines dispensed'.
2. **Data Transformation:** data captured from the previous step is then
cleansed for duplication, quality, and potentially scrubbed for privacy poli-
cies. This data then goes through a multifaceted enrichment process to
140 facilitate data analysis. For instance, a journey of the veterinary nurse
can be captured at every stage, enriched with demographics and animal
breed for regression analysis and aggregate views.
3. **Data Serving:** at this stage, data is ready to be served to diverse array of
needs ranging from machine learning to marketing analytics, to business
145 intelligence to product analysis and customer journey optimization. In the
'veterinary practice management software' the data platform can provide
real-time data through event backbone system such as Kafka about cus-
tomers who have applied and have been dispensed Restricted Veterinary
Medicine (RVM) to make sure that these transactions comply with the

150 conditions of the registration of these products.

The lifecycle depicted is indeed a high-level abstract view of prevalent BD systems. Howbeit, it highlights an important matter; these systems are all operating underlying monolithic data pipeline architecture that tends to account for all sorts of data. This means, data that logically belong to different domains
155 are now all lumped together and crunched in one architectural construct, making maintainability and scalability a daunting task [20].

While architectures in software engineering have gone through series of evolution in the industry, adopting a more decentralized and distributed approaches such as microservice architecture, event driven architectures, reactive systems,
160 and domain driven design [21], the data engineering, and in specific BD ecosystems do not seem to be adopting many of these patterns. Evidence collected from this study have proven that attention to decentralized BD systems, meta-data, and privacy is deficient. Therefore, the whole idea of 'monolithic data pipeline architecture with no clearly defined domains and ownership' brings significant challenges to design, implementation, maintenance and scaling of BD
165 systems.

To address these issues, we explore a domain-driven distributed RA for BD systems and propose an RA that addresses some of the challenges. The RA is inspired by the advances in software engineering architectures, and in specific
170 microservices, domain-driven design, and reactive systems.

2.2. *Why reference architecture?*

To justify why we have chosen reference architectures as the suitable artefact, first we have to clarify two assumptions;

1. having a sound software architecture is essential to the successful development and maintenance of software systems [22]
175
2. there exist a sufficient body of knowledge in the field of software architecture to support the development of an effective RA [1]

One of the focal tenets of software architecture is that every system is developed to satisfy a business objective, and that the architecture of the system is
180 a bridge between abstract business goals to concrete final solutions [22]. While the journey of BD can be quite challenging, the good news is that a software RA can be designed, analyzed and documented incorporating best practices, known techniques, and patterns that will support the achievement of the business goals. In this way, the complexity can be absorbed, and made tractable.

185 Practitioners of complex systems, software engineers, and system designers have been frequently using reference architectures to have a collective understanding of system components, functionalities, data-flows and patterns which shape the overall qualities of system and help further adjust it to the business objectives [23]; [24]. There is a fair amount of literature on reference architectures, and whereas different authors definition may vary, they all share the same
190 tenets.

A reference architecture is amalgamation of architectural patterns, standards, and software engineering techniques that bridge the problem domain to a class of solutions. This artefact can be partially or completely instantiated
195 and prototyped in a particular business context together with other supporting artefact to enable its use. RAs are often created from previous RAs [1].

The usage of RAs for the development of complex systems is not new. In software product line (SPL) development, RAs are generic artifacts that are configured and instantiated for a particular domain of systems [25]. In software
200 engineering, major IT giants like IBM has referred to RAs as the 'best of best practices' to address unique and complex system development challenges [23].

Based on the premises discussed and taking all into consideration, RAs can facilitate the issues of BD architecture and data engineering because of the following reasons;

- 205 1. RAs can promote adherence to best practice, standards, specifications and patterns
2. RAs can endow the data architecture team with openness and increase

operability, incorporating architectural patterns that ensue desirable pre-defined quality attributes

- 210 3. RAs can be the best initial start to the BD journey, capturing design issues when they are still cheap
4. RAs can bring different stakeholders on the same table and help achieve consensus around major technological constructs
5. RAs can be effective in identifying and addressing cross-cutting concerns
- 215 6. RAs can serve as the organizational memory around design decisions, enlightening next subsequent decisions
7. RAs can act as a summary and blueprint in the portfolio of software engineers and software architects, resulting in better dissemination of knowledge

220 3. Research Methodology

Our research methodology is made up of two major phases. First we explore the body of knowledge in academia and industry to identify architecturally significant requirements (ASR) for BD systems, and secondly we discuss the chosen methodology for developing the artifact

225 3.1. Requirement Specification

Architecture aims to produce systems that are addressing specific requirements, and one cannot succeed in designing a successful architecture if requirements are unknown [22]. Therefore, in this section, we strive to firmly define the requirements necessary for the development of Cybermycelium. We present 230 with four integral pieces of information:

1. Determining the type of the requirement
2. Identifying the right approach to categorization of the requirements
3. Identifying a rigorous approach for presentation of the requirements

For maximum clarity, we've mapped the following sub-sections against the 235 above mentioned elements.

3.1.1. *Type of requirements*

Precursor to theorizing about the potential of Cybermycelium, we needed to define what are the requirements. System and software requirements come in different flavours and can range from a sketch on a napkin to formal (mathematical) specifications. Therefore, we first needed to identify what kind of requirements is the most suitable for the purposes of this study. To answer this question, we first explored the body of evidence to understand the current classification of software requirements.

There's been various attempts to defining and classifying software and system requirements. For instance, Sommerville [26] classified requirements into three levels of abstraction that are namely 1) user requirements, 2) system requirements and 3) design specifications. The author then mapped these requirements against user acceptance testing, integration testing and unit testing. While this could satisfy the requirements of this study, we opted for a more general framework provided by Laplante [27]. In Laplante's approach, requirements are categorized into three categories of 1) functional requirements, 2) non-functional requirements, and 3) domain requirements.

Our objective is to define the high-level requirements of big data systems, thus we do not fully explore 'non-functional' requirements. Majority of non-functional requirements are emerged from the particularities of an environment, such as a banking sector or healthcare, and do not correlate to our study. Therefore, our primary focus is one functional and domain requirements and secondly on non-functional requirements.

3.1.2. *Categorizing requirements*

After having filtered out the right type of requirement, we then sought for a rigorous and relevant method to categorize the requirements. For this purpose, we followed the well-established categorization method based on BD characteristics, that is the 5Vs. These 5Vs are velocity, veracity, volume, Variety and Value [28], [29]. Nadal et al. [30] have underpinned their reference architecture on these characteristics and requirements that goes with them. Moreover, NIST

BD Public Working Group embarked on a large scale study to extract requirements from variety of application domains such as Healthcare, Life Sciences, Commercial, Energy, Government, and Defense. The result of this study was the formation of general requirements under seven categories. In another effort
 270 by Volk et al. ([31]), nine use cases for BD projects are identified by collecting theories and use cases from the literature and categorizing them using a hierarchical clustering algorithm. Bashari et al. ([32]) focused on the security and privacy requirements of BD systems, Yu et al. presented the modern components of BD systems ([33]), Eridaputra et al. ([34]) created a generic model for
 275 BD requirements using goal oriented approaches, and Al-jaroodi et al. ([35]) investigated general requirements to support BD software development.

We've also studied other reference architectures developed for BD systems to understand general requirements. In one study, Ataei et al. [36] assessed the body of evidence and presented with a comprehensive list of BD reference
 280 architectures. This study helped us realize the spectrum of BD reference architectures, how they are designed and the general set of requirements. By analyzing these studies and by evaluating the design and requirement engineering required for BD reference architectures, we adjusted our initial categories of requirements and added security and privacy to it.

285 3.1.3. *Present requirements*

After knowing the type and category of requirements, We looked for a rigorous approach to present these requirements. There are numerous approaches used for software and system requirement representation including informal, semiformal and formal methods. For the purposes of this study, we opted for
 290 an informal method because it is a well established method in the industry and academia [37]. Our approach follows the guidelines explained in ISO/IEC/IEEE standard 29148 [38] for representing functional requirements. We have also taken inspiration from Software Engineering Body of Knowledge [39]. However, our requirement representation is organized in term of BD characteristics.

Veracity	<p>Ver-1) System needs to support data quality curation including classification, pre-processing, format, reduction, and transformation,</p> <p>Ver-2) System needs to support data provenance including data life cycle management and long-term preservation.</p>
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295 3.2. The artifact development methodology

There are a few studies that have addressed the systematic development of reference architectures. Cloutier et al [23] present a high-level model for RA development through collection of contemporary information and capturing the essence of architectural advancements. In another effort, PuLSE-DSSA is proposed by Bayer et al. [40] in the context of product line development and domain engineering. PuLSE-DSSA emphasizes on capturing the existing architectural knowledge. Stricker et al. [41] propose a pattern-based approach for creating an RA. This study revolves around software engineering patterns motivated by the work of Gamma et al [42]; and proposes a structural approach that includes three layers of patterns with well-defined hierarchical relationships. Nakagawa, Martins, Felizardo, and Maldonado [43] propose an approach to RA design outside of product line management context that is concentrated towards aspect-oriented systems.

Galster and Avgeriou [44] propose an empirically grounded reference architecture based on two main facets; Existing RAs in practice and available literature on RAs. Along the same vein, Nakagawa et al [45] presented ProSA-RA which is a 4 phase methodology that unlike many other methodologies do provide a more comprehensive instructions on RA evaluation. In addition, this methodology benefits from an ecosystem of complementary constructs that aid in RA design and evaluation such as RAModel [46] and a framework for evaluation of RAs (FERA) [47]. In a recent study, Derras et al. [48] propose a schema of practical RA development in the context of software product line and domain engineering. This study is based on capturing knowledge from architectures in practice with attention to variability, configurability and product line devel-

opment. The findings provide a four-phase process to develop quality driven reference architectures. This approach is influenced by ISO/IEC 26550 [49].

By analysis and study of all these approaches for design and development of RAs, a common pattern has been witnessed. Whereas some of them are more recent and some belong to years ago, there are commonalities that has been observed. All these approaches are grounded on three main pillars, 1) Existing RAs 2) RAs in literature 3) Architectures in practice. Taking this into consideration and by analyzing the results of the systematic literature review conducted by Ataei et al [1] we found 'Empirically-grounded reference architectures' proposed by Galster and Avgeriou [44], a suitable methodology, because firstly it's been adopted by many studies, and secondly it's comparatively in-line with the nature of our study.

Nevertheless, we did not fully adopt this methodology and rather customized to the needs of this particular research. This is due to some inherent limitations that has been witnessed with the methodology. For instance we could not find a comprehensive guideline on how to identify data sources and how it could be categorized and synthesized into the creation of the RA in the third step of the methodology, therefore we employed the Nakagawa's information source investigation guidelines and the overall idea of the RAModel. Another limitation we've faced was with evaluation of the RA. As evaluation, second to a sound research methodology is one of the key elements of any good design science research, we had to look for a stronger and more systematic evaluation approach than what was discussed in 'empirically grounded RAs' methodology. For this purpose, and inspired by the works of Angelov et al [50]; [51], we first created an prototype of the RA in practice, and then used 'The architecture tradeoff analysis method' (ATAM) [52] to evaluate the artefact.

This research methodology is constituent of 6 phases which are respectively; 1) Decision on the type of the RA 2) Design strategy 3) Empirical acquisition of data 4) Construction of the RA 5) Enable RA with variability 6) Evaluation of the RA. The phrase 'empirically grounded' refers to two major elements; firstly the reference architecture should be grounded in well-established and

proven principles; secondly, the reference architecture should be evaluated for applicability and validity. These don't only belong to Galster and Avgeriou methodology, and other researchers such as Cloutier [23] and Derras et al [25] have promoted the same ideas.

355 It is worth mentioning that this methodology is iterative, meaning that the results gained from the evaluation phase (6th phase) determines the subsequent iterations until the design reaches saturation.

3.3. Step1: Decision on type of the RA

Precursor to any effective RA development, is the decision on type of it. The
360 type of the RA is significant, as it illuminates on information to be collected and the construction of the RA in later phases. The selection on the type of RA for the purposes of this study is based on two dimensions; the classification framework proposed by Angelov et al. [53] and the usage context [54].

Based on the classification framework proposed by Angelov et al. [53], five
365 types of RA are defined. This framework has been developed with the goal of supporting analysis of RAs with regards to context, goal, and the architecture specification/design relationships. It is based on 3 major dimensions namely context, goals, and design, each having their own corresponding sub-dimensions. These dimensions and sub-dimensions are derived by the means of interrogatives
370 (the usage of interrogates is a well-established practice for problem analysis).

The interrogatives 'When', 'Where', and 'Who' have been used to address the 'context', 'Why' has been used to address 'goal', and 'How' and 'What' have been used to address 'design' dimension. The outcome of the study categorizes RAs in two major groups; 1) standardization RAs and 2) Facilitation RAs. This
375 framework has been chosen because it is completely in-line with the purposes of this study and aims to demarcate a clear domain for the RA to be developed. The comprehensive classification of the RAs with examples in practice illuminates on how different RAs are playing roles in the industry and how they are classified. This brings clarity on what should be developed and what boundaries
380 should be drawn.

By reading the results of the recent SLR conducted by Ataei et al on BD RAs [1], we've added more examples of the RAs on top of what was provided by Angelov [53], and provided the following updated list of RA classifications with examples;

- 385 1. Standardization RAs
 - (a) Type 1: classical, standardization architectures designed to be implemented in multiple organizations. Examples are:
 - i. WRM [55]
 - ii. OSI RM [56]
 - 390 iii. OATH [57]
 - iv. COBRA [58]
 - v. Neomycelia [3]
 - vi. Kappa [59]
 - vii. Bolster [14]
 - 395 (b) Type 2: classical, standardization architectures designed to be implemented in a single organization
 - i. Fortis Bank Reference Software Architecture [53]
- 2. Facilitation RAs
 - (a) Type 3: classical, facilitation reference architectures for multiple or-
400 ganizations designed by a software organization in cooperation with user organizations
 - i. Microsoft Application Architecture for .Net [60]
 - ii. IBM PanDOORA
 - iii. OATH [57]
 - 405 iv. COBRA [58]
 - (b) Type 4: classical, facilitation architectures designed to be implemented in a single organization
 - i. Achmea Software Reference Architecture [61]
 - ii. ABN-AMRO Web Application Architecture [62]

- 410 (c) Type 5: preliminary, facilitation architectures designed to be implemented in multiple organizations
- i. ERA [54]
 - ii. AHA [63]
 - iii. eSRA [64]

415 The domain driven distributed BD RA chosen for the purposes of this study pursues two major goals; 1) enabling and support the development and data engineering of BD systems 2) concurrently ensuring that interoperability between different heterogeneous components of the BD system is established. Therefore, the outcome artefact will be a BD RA that is a classical standardization RA
420 designed to be implemented in multiple organizations.

3.4. Step2: Selection of Design Strategy

Angelov et al [50] and Galster et al[44] have both presented that RAs can have two major design strategies to them; 1) RAs that are designed from scratch (practice driven), 2) RAs that are based on other RAs (research driven). Designing RAs from scratch is rare, and usually takes place in an emergent domain
425 that have not perceived a lot of attention. On the other hand, most RAs today are the amalgamation of a priori concrete architectures, models, patterns, best practices, and RAs, that together provide a compelling artefact for a class of problems.

430 RAs developed from scratch tend to create more prescriptive theories, whereas RAs developed based on available body of knowledge tends to provide with more descriptive design theories. The RA designed for the purposes of this study is a research-based RA based on existing RAs, concrete architectures, and best practices.

435 3.5. Step 3: Empirical Acquisition of Data

As aforementioned, due to the limitation witnessed by this research methodology, we have augmented this phase, and increase the systematicity and transparency of data collection and synthesis through various academic methods such as systematic literature review or SLR.

440 This phase is made up of three major undertakings; 1) identification of data sources; 2) capturing data sources; 3) synthesis of data sources.

3.5.1. Identification of data sources

To identify suitable data sources, we've employed the first step of ProSA-RA methodology, 'information source investigation'. This step is an endeavour
445 to capture focal and ancillary knowledge and theories that revolve around the target domain, and lay the ground of RA development.

To unearth the architectural quanta, and to highlight gradations between various approaches to BD system development, we've selected most relevant sources as the followings;

- 450 1. **Practice-led conferences:** given that majority of recent advancements for emerging technologies such as microservices architecture [65];[66] [67] and BD are coming from virtually hosted practice-led conferences, we've chosen some of the best conferences hold world-wide for the purposes of data collection. These conferences are 1) Qcon [68] 2) State of Data Mesh
455 by ThoughtWorks [69] 3) Worldwide Software Architecture Summit'21 [70] and 4) Kafka Summit Europe 2021 [71]. Our objective was to capture the frontiers of software architecture and emerging approaches currently being practiced in IT giants such as Google, Facebook and Netflix. Among all the speech in these conferences, we looked for topics that entailed or
460 were related to the keywords 'emergent software architecture trends', 'distributed software architecture', and 'BD software architecture'. We used the software Nvivo to code the transcripts from the conference videos. We used the aforementioned keywords as the codes and associated different texts, summative, essence-capturing sentences, evocative attributes to
465 them. During this process, a new theme 'domain driven design' emerged. We added that into the list of codes as well.
2. **Publications:** in order to capture evidence from the body of knowledge, we conducted a systematic literature review (SLR), following the guidelines of PRISMA presented by Moher et al [72], and Kitchenham et al.

470 [73]. Although PRISMA is a comprehensive guidelines for conducting a
systematic literature review, it is derived from the healthcare commu-
nity and is driven by assumptions that may not be thoroughly relevant
to software engineering and information system researchers. To this end,
we adopted some guidelines from Kitchenham et al. for evidence based
475 software engineering.

The main objective of this SLR was to highlight common architectural
constructs found among all the BD RAs. This SLR is build on top of our
recent work [1] that covered all the RAs by 2020.

The initial SLR included IEEE Explore, ScienceDirect, SpringerLink, ACM
480 library, MIS Quarterly, Elsevier, AISel as well as citation databases such
as Scopus, Web of Science, Google Scholar, and Research Gate. The SLR
search keywords used were 'Big Data Reference Architectures', 'Reference
Architectures in the domain of Big Data', and 'Reference Architectures
and Big Data'. We followed the same methodology, but this time for the
485 years 2021 and 2022. Our aim was to find out if there has been any new
BD RA published during the years mentioned.

By the result of this SLR, we've found 3 more BD RAs ([3]; [74]; [75])
and we've added two new standards ([76]; [77]) to further solidify our
study. Converging these new SLR with the old, covering the years 2010-
490 2022, we've pooled 89 literature in the primary phase, and another 10 by
snowballing and citation searching. These 99 literature then went through
our inclusion, exclusion and quality criteria. These criteria is as blow;

- Inclusion criteria:

- (a) Primary and secondary studies between Jan 1st 2020 and Sep
495 1st 2022 on the topics of BD RA, BD architecture, and BD ar-
chitectural components were included
- (b) Research that Indicates the current state of RAs in the field of
BD and demonstrates possible outcomes
- (c) Studies that are scholarly publications, books, book chapters,
500 thesis, dissertations, or conference proceedings

(d) Grey literature such as white paper that includes extensive information on BD RAs

- Exclusion criteria:

(a) Informal literature surveys without any clearly defined research questions or research process

(b) Duplicate reports of the same study (a conference and journal version of the same paper)

(c) Short papers (less than 5 pages)

(d) Studies that are not written in English

This screening process was conducted by each researcher separately. Disagreement among researchers were resolved using Krippendorff's alpha ([78]). Our aim was not to get involved in a very complicated statistics model, so we've done most of the computations using SPSS, specifically with Hayes' Macro. Our α value was within the acceptable range (above 80).

After excluding papers based on inclusion and exclusion criteria, and as suggested by Kitchenham et al [73], we assessed studies based on their quality. For this purposes, and inspired by Critical Appraisal Skills Programme (CASP [?]), and JBI's critical appraisal tool ([?]), we developed our quality framework based on 7 criteria. The 7 criteria tested literature on 4 major areas that can critically affect the quality of the studies. These categories and the corresponding criteria are as following;

1. *Minimum quality threshold:*

(a) Does the study report empirical research or is it merely a 'lesson learnt' report based on expert opinion ?

(b) The objectives and aims of the study is clearly communicated, including the reasoning for why the study was undertaken ?

(c) Does the study provide with adequate information regarding the context in which the research was carried out ?

2. *Rigour:*

- 530 (a) Is the research design appropriate to address the objectives of the
research ?
- (b) Is there any data collection method used and is it appropriate ?
3. *Credibility:*
- (a) Does the study report findings in a clear and unbiased manner ?
- 535 4. *Relevance:*
- (a) Does the study provides value for practice or research

Taken all together, these 7 criteria gave us a measure of the extent to which a particular study's findings could make a valuable contribution to the review

3.5.2. *Data Synthesis*

540 After pooling the studies, we removed 11 studies before screening because either they were duplicates or they were not in English. The remaining 78 studies went through screening, in which, 2 studies excluded based on the exclusion criteria. From there on, 76 studies have been assessed for eligibility based on the quality framework and the inclusion criteria. The result of this process handed
545 over 67 studies from this branch. From the other branch, 10 records identified through citation searching. These reports have been assessed through the same quality framework, inclusion and exclusion criteria, which yielded 5 studies from this stream. Together 68 studies pooled for this SLR as depicted in figure 1.

These 68 studies are comprising of journal papers, conference papers, book
550 chapters, tech reports, tech surveys white papers, standards, master dissertation, and PhD thesis. Out of the pool of these studies, 39.4% are from IEEE Explore, 4.4% are from ScienceDirect, 23.5% are from Springerlink, 13.2% are from ACM, and 29.4% are from other sources such as citation search, Google Scholar and Research Gate. 30 journal articles, 14 conference papers, 6 whitepapers,
555 2 ISO standards, 14 book chapters, and 2 postgraduate studies have been selected. 26% of these studies are from the year 2010-2013, 33% are from the years 2013-2015, and 51% are from the years 2016-2022. These stats are portrayed in figure 2.

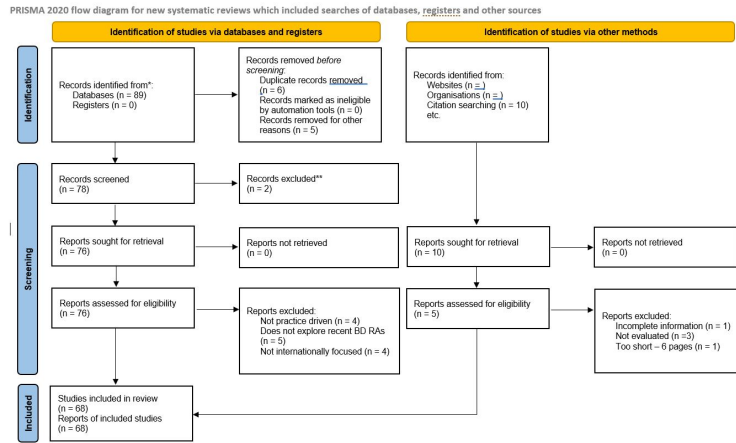


Figure 1: PRISMA flowchart

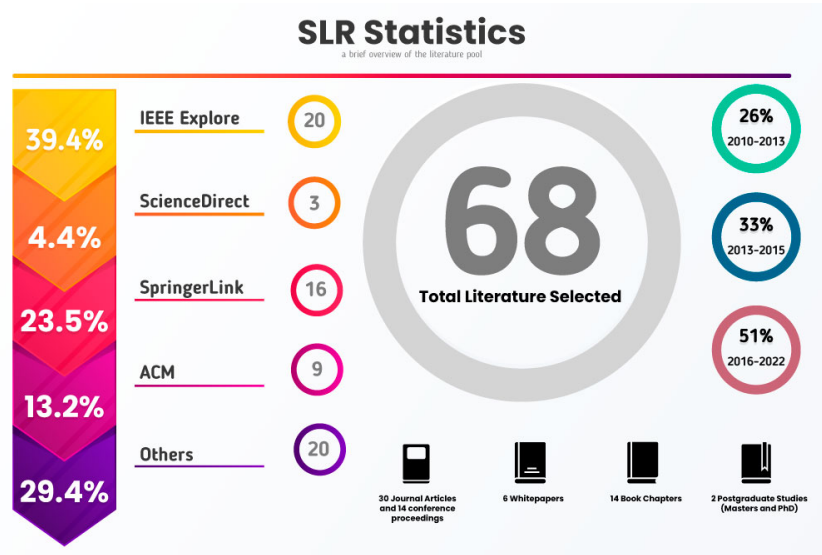


Figure 2: SLR statistics

By this stage, the research objective is set, studies are pooled, assessed and refined, thus the research embarked on the actual synthesis of data. For this purpose, we employed thematic synthesis presented by Cruzes et al. [79]. An integral element of this phase is data extraction, in which the essence of the studies are obtained in an explicit and consistent manner. We opted for an

integrated approach to coding ([80]) using the software Nvivo Nvivo [81]. All the
565 keywords aforementioned has been created as nodes in the software, which are
then associated to relevant sentences in studies. After coding all the studies, the
findings have been synthesized to create theories, which in turn emerged themes
and patterns. The findings gained from this SLR grounded the foundation for
various aspect of the SLR development. To increase transparency, RAs and
570 standards found as the result of this SRL is presented in Appendix A.

3.6. Construction of the RA

Based on the themes, theories, and patterns realized in the previous steps,
the process of RA construction took place. Integral to this step was the iden-
tification of elements that the RA should contain, how these elements should
575 be synthesized, and how the RA can be portrayed and communicated. To de-
scribe our RA, we followed ISO/IEC/IEEE 42010 standard [82]. This standard
pivots on concrete architectures, so we did not 100% conform to it, but rather
the good and relevant parts of it has been taken. For instance, architecture
viewpoints, statement of corresponding rules, and expression of the architecture
580 through architecture description languages (ADLs) have had direct impact on
the construction of this RA.

A key challenge in the development of this RA was to strike a balance be-
tween the specificity of the micro patterns and approaches to system develop-
ment and general architectural concepts that reflect a view of the system as an
585 array of interrelated entities. Angelove et al [83] approached this problem by
the means of interrogative through a defined framework that aims to guide the
creation of RAs. Cloutier et al [23] suggest that a RA should entail technical,
business and customer context views, whereas Vogel et al [84] provided classi-
fies RA views based on the usage context, as industry specific, platform specific,
590 industry crosscutting and product line RAs.

Stricker et al [85] expressed their pattern-based RA by adhering several
distinct views into one. Chang et al [86] presented NIST BD RA as a system
constituent of logical components connected though interoperability interfaces

in several fabrics. On the other hand, ISO/IEC/IEEE 42010 refrains from using
595 phrases such as “technical architecture”, “physical architecture”, or “business
architecture”.

Taking the best evidence from the available body of knowledge, We decided
to adhere several views into one and express the RA through a multi-layer
modeling language called Archimate. Archimate is a mature modeling lan-
600 guage developed by Open Group that provides with a uniform representation of
high-level architectural diagram aimed at portraying and delineating enterprise
architecture [87]. Archimate being listed as a standard architecture description
language in ISO/IEC/IEEE 42010, is designed based on a set of related con-
cepts that are specialized towards the system at different architectural layers.
605 This means that the architect is enhanced with an integrated architectural tool
that visualizes and describes different architecture domains and their underlying
relations [88]; [89].

Archimate utilizes service-orientation to distinguish and relate the applica-
tion, business and technology layer and use realization relationships to create
610 relationship between concrete elements and more abstract elements across three
layers. In addition, Archimate can be customized to account for varying needs
of the architect.

3.7. Enabling RA with variability

Enabling RA with variability is an important process that helps with the
615 instantiation of it. This allows RA to remain useful as a priori artefact when it
comes to organization-specific regulations, and regional policies that constrain
the architect design decisions [90].

Variability management has been studied in the domain of Business Pro-
cess Management (BPM) [91]; [92]; [93] and Software Product Line Engineer-
620 ing (SPLE) [94][95]; [96]; [97]; [98]. In BPM, variability management revolves
around efficient handling of different variants in business processes, whereas in
SPLE, variability management is about modifying and extending the software
artefact to account of the requirements of a specific context.

Clear identification of variability and explicit communication of it improves
625 communication between stakeholders, allows for traceability between variation
causes and effects and facilitates the decision making [99].

Variation points are decided based on the data collected in previous steps.
Galster et al [44] suggest that there are three approaches to enabling variability;

1. Annotation of the RA
- 630 2. Variability views
3. Variability models

We could not find an in-detail explanation of how one should choose the
appropriate variability enabling approach. Therefore, inspired by the works of
Rurua et al [90], we decided to extend the RA with variability, by the means of
635 Archimate annotations. We have achieved this in two steps; first we developed
a custom layer that represents focal variability concepts, and then we extended
the RA through annotation. The aim of this process is not find all variability
points that may emerge in the usage context, but to provide with high-level
system related architectural variabilities that an architect may consider for im-
640 provement of design and adoption of the RA.

The variability model is depicted in Fig 3 by the means of Archimate's
motivation layer. This modeling is driven by the works of Pohl et al [94] and in
specific their graphical notation of variability information, and Rurua et al [90]
and in specific, their variability management concepts model.

645 3.8. *Evaluation of the RA*

Evaluation of the RA is to ensure that it has achieved the goals stated prior
to development, to test its effectiveness and usability, and to make sure that it
addresses the identified problems. Two fundamental pillars of the evaluation is
the correctness and the utility of the RA and how efficient it can be adapted
650 and instantiated [44]. The quality of RA can be assessed by how it can be
transformed into an effective organization-specific concrete architecture. The
fact that this RA is built upon former RAs helps making the evaluation steps

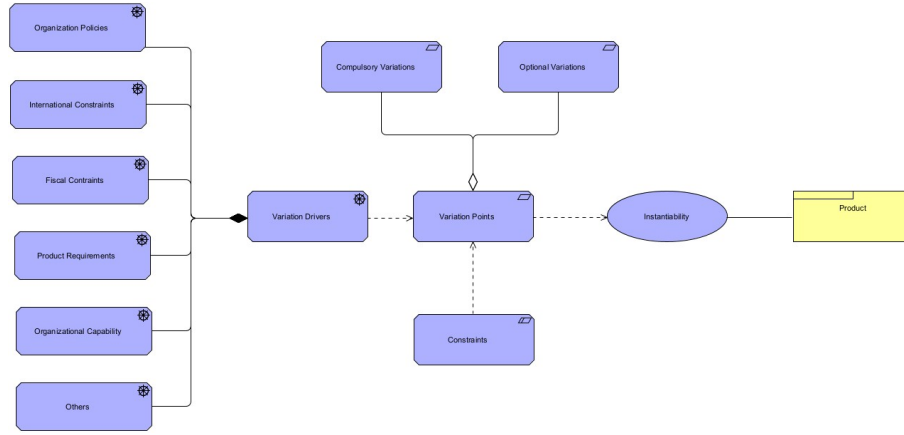


Figure 3: Variability management concepts model

easier as the research can get inspiration from other studies and their approach to evaluation [100].

Nevertheless, evaluation of the RAs is a well-known challenge among researchers [54]; [101]; [102]; [103]. RAs and concrete architectures have distinct qualities. They vary in at least 3 major ways;

1. RAs are of higher level of abstraction
2. in RAs stakeholders are not clearly grouped
3. RAs tend to be focused more on architectural qualities

While there are many well-established methods for assessing concrete architectures such as Scenario-based Architecture Analysis Method [104], Architecture Level Modifiability Analysis [105], Performance Assessment of Software Architecture [106], Architecture Trade-off Analysis Method [107], none of these methods can be directly applied to evaluate RAs. To support this statement, three major issues have been identified. These issues are as follows;

1. One of the main problems for applying existing evaluation methods to RA is the lack of clearly defined group of stakeholders [50], while ATAM and other methods are highly dependent on participation of stakeholders for evaluation. Due to the level of abstractness of the RA, reaching various

group of stakeholders and persuade them to anticipate in the study, is problematic and does not fit to the timeline of this study. Even more notably, it is unlikely that all stakeholders will unite around a common reference architecture as different members may or may not agree with the overall idea of the RAs, may come from different backgrounds, and may lack architectural visions

2. Evaluation frameworks and methods for concrete architectures make use of scenarios. Howbeit due to RAs level of abstraction, creation of usable scenario is difficult. Either a large set of scenarios should be developed covering all the aspects of the RA with regards to specific domain, or a more general scenarios should be developed to cover all the aspects. In the first approach, a large number of scenarios, makes data analysis troublesome and a tedious process. Moreover, the order of prioritization of these scenarios and defining them, and validating them is a problematic task. In the second approach, due to the generality of the scenario, evaluation of effectiveness and usability of the RA becomes difficult and may become incomplete [101]. These challenges have been observed even in the evaluation of highly complex concrete architectures in information systems domain [108].

Based on the problems discussed above, available methods of architecture analysis are not sufficient in evaluating the RA. This has been addressed by various researchers in the industry.

In one study, Angelov et al [50], modified ATAM and extended it to resonate well with RAs. This process took place by invitation of representatives from leading industries for the evaluation process, and the selection of various contexts and defined scenarios for these contexts. Furthermore, ATAM has been extended to evaluate completeness, buildability and applicability. Howbeit the selection of the right candidate and involving them in the process is a time-consuming and daunting task and may yield incomplete information. In addition, candidates maybe lacking architectural visions, increasing the threat

to validity.

In addition to extending ATAM for RAs, Graaf et al [109] presented an evaluation approach in which SAAM is extended to help reduce the organizational impact of it. In Another study by Maier et al, Maier et al. (2013) as a post-graduate thesis in Eindhoven University of Technology, the evaluation of the RA has been conducted by mapping it against existing concrete architectures described in industrial whitepapers and reports. Along the lines, Galstar et al [44] suggested reference implementations, prototyping and incremental approach for the validation of the RA.

Rohling et al [110] have evaluated their RA by mapping it against the requirements set for the study. This was facilitated by the RA research methodology created by Nakagawa et al [111] and the complementary RAModel [46].

Inspired by all the studies listed, for the purposes of this study, we will first create a prototype of the RA in an actual organizational setup and then we will use ATAM to evaluate the concrete architecture.

4. Requirement Specification

By the result of the processes conducted in 3.1, and by carefully evaluating similar approaches to requirement specification, we tailored a set of requirement for the development for Cybermycelium. These requirements are presented in Table 1. These requirements are mapped against our components discussed in 6.2.

5. Why Cybermycelium?

In this section, we present with theories that aim to explain the limitations of current BD architectures. While we have briefly discussed the current status of BD architectures and failure modes in 2.1, this section digs deeper into these challenges.

Table 1: BD system requirements

Volume	Vol-1) System needs to support asynchronous, streaming, and batch processing to collect data from centralized, distributed, and other sources, Vol-2) System needs to provide a scalable storage for massive data sets
Velocity	Vel-1) System needs to support slow, bursty, and high-throughput data transmission between data sources, Vel-2) System needs to stream data to data consumers in a timely manner, Vel-3) System needs to be able to ingest multiple, continuous, time varying data streams, Vel-4) System shall support fast search from streaming and processed data with high accuracy and relevancy, Vel-5) System should be able to process data in real-time or near real-time manner
Variety	Var-1) System needs to support data in various formats ranging from structured to semi-structured and unstructured data, Var-2) System needs to support aggregation, standardization, and normalization of data from disparate sources, Var-3) System shall support adaptations mechanisms for schema evolution, Var-4) System can provide mechanisms to automatically include new data sources
Value	Val-1) System needs to be able to handle compute-intensive analytical processing and machine learning techniques, Val-2) System needs to support two types of analytical processing: batch and streaming, Val-3) System needs to support different output file formats for different purposes, Val-4) System needs to support streaming results to the consumers
Security & Privacy	SaP-1) System needs to protect and retain privacy and security of sensitive data, SaP-2) System needs to have access control, and multi-level, policy-driven authentication on protected data and processing nodes.

5.1. *A need for a paradigm shift*

If our aspiration to enhance every business aspect with data needs to come to fruition, we need a different approach to data architecture. Traditional data warehouse approaches to business intelligence, while have addressed the volume and computing aspect of data, have failed to address other characteristics of it; heterogeneity and proliferation of data sources (variety), the speed at which data arrives and needs to be processed (velocity), the rate at which data mutates (variability), and the truth or quality of the data (veracity).

Suppose company A would want to adopt a BD initiative to embark on this complex endeavour, what would be the first step ? does the company have to worry about high-throughput stream processing ? does it have to worry about regional privacy regulations? does it have to worry about cost-efficient batch processing? perhaps yes to all of these questions, but what's even more integral to the success of the whole endeavour is the underlying architecture that governs the entire system, its components, their relations to each other, data flow and principles and standards that govern the quality attributes and evolution of the system.

This architecture and design process if done underlying current prevalent approaches, can bring about colossal losses, and leave many managers disappointed. Nevertheless, we don't claim that all these architectures will fail, perhaps some have proven to be successful in a specific context. There are two threats to maintainability and scalability of these systems;

1. **Data source proliferation:** as the BD system grows and more data sources are added, the ability to ingest, process, and harmonize all these data in one place diminishes. This in turn, reduces maintainability, makes company reliant on lead data engineers who built the infrastructure, and makes scaling these systems very difficult. The proliferation of data sources if not managed carefully, can result in data swamps as well, which makes understanding data domains, and providing data as a service a complicated task.

2. **Data consumer proliferation:** organizations that utilize rapid experimentation approaches such Hypothesis-Driven Development and Continuous Delivery [112] constantly introduce new use cases for data to be consumed in different domains. This means that variability of the data rises, and the sum of aggregations, projections, and slices increases, which in turn adds more work to the backlog of the data engineering team, slowing down the process of serving the data to consumers. Inability to account for the data consumer demands can be a point of friction in organizations [20].

To address these challenges, the lead architect or the architecture governance group will then have to choose the right architectural quanta to segregate the monolith. According to Ford et al [113], an architectural quantum is a component of a system with high functional cohesion that is independently deployable, this is also referred to as a service [114]. The main motivation for segregating the monolith into its architectural quantum, is to parallelize work in various business domain, to reach a better velocity, reduce cost, promote ownership, increase performance and reach higher operational scalability.

Currently, these architectures are usually segregated into pipelines that each process data differently. While each pipeline has its own responsibility to handle various aspect of the BD system, there is still a high coupling between the pipelines, as 'data cleansing' phase cannot start after 'data ingestion'. This coupling is even more highlighted when the company is at the stage of rapid experimentation with data sources and would like to explore new domains of insight generation, and this in turn means that delivering new features and values is orthogonal to the axis of change.

Using the same practice management software example, given that a new class of animals (equine animals as opposed to small animals) should be incorporated for data analysis; the data engineering team should then modify and extend the whole pipeline of ingest, process and serve to account for the particularity of the data captured regarding this new class of animals. New ingestion

services required, the schema might change, the veracity checking mechanisms might differ, cleansing varies and more. This implies an end-to-end dependency which affects external teams, slows down processes and make maintenance gradually harder. This implies that using pipeline as an architectural quantum in such a coupled way is perhaps not the most efficient architecture to BD systems.

Another major issue with the current architectural approaches is that data engineering is usually confined into a team of hyper-specialized individuals who are siloed from the operational units of the organization. These teams, being fully responsible for creating the infrastructure for data processing, are often absent of business knowledge and the domain, which limits their productivity. These individual usually have a limited understanding of the data sources, data provenance, data consumers, the changing nature of the business domains, the overall product vision, and the application side of things, yet they are responsible to provide data for a large array of analytical and operational needs in a timely manner. For instance, given a context in which a product owner and application developer can cooperate, the synergy between the data engineer, application developer, and product owner can bring about more mature decisions that can align the requirements across various technical and logical domains and allow for various stakeholders to contend and communicate their concerns.

Whereas there could be other factors to be discussed deeply in this paper regarding current BD architectures, our aim is not to explore any further and emphasize more on the solution artefact we've designed to address some of these challenges. This artifact is discussed in the next section.

6. Cybermycelium: A Domain Driven Distributed Reference Architecture for Big Data Systems

This section is constituent of two integral elements: theory and artifact. First we begin by exploring the major architectural constructs that underpins our artifact and then we present the artifact and describe its components.

815 6.1. The Theory

There are various design and kernel theories employed to justify our artifact and the decisions made. These theories are described in following sub-sections.

6.1.1. A paradigm shift: distributed domain driven architecture

Based on the premises discussed in 5.1, one can infer that the idea of monolithic and centralized data pipelines that are highly coupled and operated by silos
820 of hyper-specialized BD engineers has limitations and can bring organizations into a bottleneck .

We therefore, explore a domain driven distributed and decentralized architecture for BD systems and posit that this architecture can address some of
825 the challenges discussed. This idea is inspired by the advancements in software engineering architecture, and in specific event-driven microservice architecture [115], domain driven design [116], and reactive systems [117].

Data usually comes into two different flavours; 1) operational data which serves the need of an application, facilitates logic, and can include transactional
830 data and 2) analytical data which usually has the temporality to it, and is aggregated to provide with insights.

These two different flavours, despite being related, have different characteristics and trying to lump them together may result in a morass. To this end, Cybermycelium realizes the varying nature between these two planes and re-
835 spects the difference. Cybermycelium aims to transfigure current architectural approaches by proposing an inversion of control, and a topology based on product domains and not the technology [118]. Our proposition is that handling two different archetypes of data, should not necessarily result in siloed teams, heavy backlogs, and a coupled implementation.

840 To further elucidate on this matter, we take the example of the microservice architecture. As the industry sailed away from the monolithic n-tier architectures into Service Oriented Architectures (SOA), organizations faced a lot of challenges. One prevalent issue was around the maintenances of Enterprise Service Bus (ESB) or SOA bus, which is the locus of aggregation. While the

845 aggregation layer could be written very thin, the reality is that the transformation of XML and logical operations started to bloat the SOA bus. This added a new level of coupling between internal and external elements of the system as a whole [119]; [120]; [121].

Microservices architecture, being the evolution of SOA, move away from
850 smart pipelines into dumb pipelines and smart services removing the need for the locus of aggregation and control. Moreover, there was no business logic written in the pipelines, and each service was segregated usually with the help of domain driven design.

Whereas microservices architecture still have its challenges, the gradations
855 of software architectures in software engineering industry can be analogous to the data engineering domain. One can perceive the pipeline architecture and its coupling nature similar to SOA and the business logic written to connect the services in the SOA bus.

Based on these premises we posit 4 underpinning principles for Cybermycelium;

- 860
1. Distributed Domain driven services with bounded context
 2. Data as a service
 3. Data infrastructure automation
 4. Governance through a federation service
 5. Event driven services

865 *6.1.2. Distributed Domain driven services with bounded context*

Integral to Cybermycelium, is the distribution and decentralization of services into domains that have clear bounded context. Perhaps one the most challenging things one might face when it comes to architecting a distributed system is: based on what architectural quanta should we break down the system.
870 This issue has been repeatedly discussed for example among adopters of microservices architecture . Cybermycelium, inspired by the concept of domain-drive design, tends to sit data close to the product domain that relates to it. This implies that data inheres in the product domain and as a facet of it [66].

This is mainly driven by the fact that most organizations today are decomposed based on their products. These products are the capability of the business that are segregated into various domains. Domain's bounded context is operated by various teams with different visions and concerns, incorporating data into a bounded context can result in a synergy that can improve the management of evolution and continuous change. This can be micro, such as application developers communicating with data engineers about collecting user data in a nested data structures or in flat ones, or macro, such as application developers thinking about redesigning their graphql schema in an intermediary layer that may affect the data engineers ingestion services.

It is worth mentioning that, we are absorbing the concept of domain-driven design into this study to facilitate communication and increase the adoption, rigour and relevance of our RA. Communication is a key component of any software development endeavour [122], and without it essential knowledge sharing can be compromised. Often data engineers and business stakeholders have no direct interaction with one another. Instead, the domain knowledge is translated through intermediaries such as business analyst or project managers to series of tasks to be done [123]. This implies at least two translations from two different ontologies.

In each translation, information is lost, which is the essential domain knowledge, and this implies risk to the overall data quality. In such data engineering process, the requirement often gets distorted, and data engineer has no awareness of the actual business domain and the problem being addressed. Often times, problems being solved through data engineering, are not simple mathematical problems or a riddle, but rather have broader scopes. An organization may decide to optimize workflows and process through continuous data-driven decision making, and a data architecture that is overly centralized and not flexible can risk a project failure.

To this challenge, domain-driven design proposes a better approach to convey knowledge from domain experts to data engineers. In domain-driven design, instead of intermediary translations business domains are projected into actual

905 software engineering, emphasizing on creation of one shared terminology, that is the ubiquitous language. We do not aim to explore all facets of domain-driven design in this study, but it's worth mentioning that each business has it's own domain, and constituent core, generic and supporting sub-domains, and this is defined at the moment of RA instantiation.

910 *6.1.3. Data as a service:*

Data can be conceived as the fourth dimension of a product next to UI/UX, business and application (6.1.3). Each domain provides its data as a service. This data is consisting of both operational and analytical data. This also implies that any friction and coupling between data is removed. For instance, the
915 'invoice' domain will provide the transactional data about number of invoices and total of discounts with analytical data such as which practices have created what number of invoices in what period of time.

Product Domain

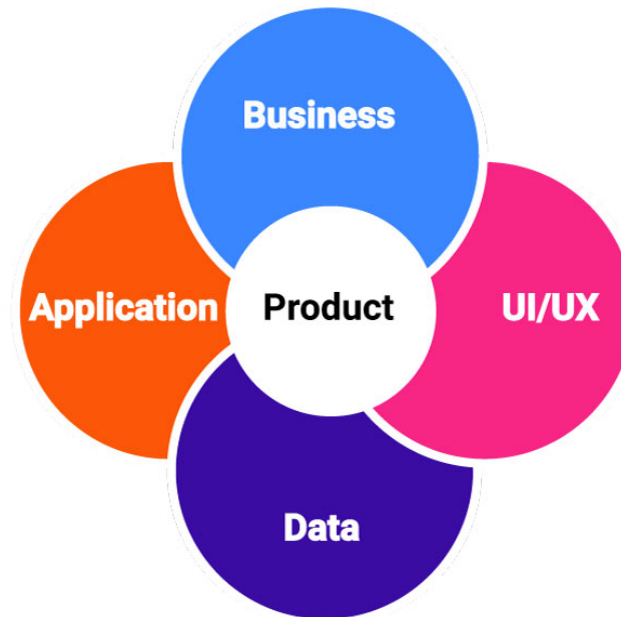


Figure 4: Product Facets

However this data-as-a-service model should be carefully implemented to account for explorability, discoverability, security, and quality. The data provided
920 as a service should have the identical qualities to customer-facing products. This also implies, that a product owner should now treat data facet as an aspect of the product and employ objective measures that assure the desired quality. These measure can be net promoter scores from data consumers, data provenance, and decreased lead time. Product owners, in addition to the application and
925 design aspect of the product, must now incorporate this new facet, and try to understand the needs of the data consumers, how they consume the data, and what are the common tools and technologies they utilized to consume the data. This knowledge can help shaping better interfaces for the product.

Product domains may also need to ingest data from upstream domains, and
930 this requires the definition of clear interfaces. Furthermore, each domain should also account for metadata. Metadata in derived from the nature of the product and its data lifecycle. Data can be ingested and served in various forms such as tables, graphs, JSON, events, and many more; but in order for the data to be useful for analytical purposes, there is a need to associate data with its
935 corresponding metadata that encompasses semantics, and history.

6.1.4. *Data infrastructure automation*

As the number of product domain increases, the effort required to build, deploy, execute, and monitor services increases. This includes the data pipelines required for that product domain to carry out its functions. The platform skills
940 required for these kinds of work is usually found in Devops engineers and site reliability engineers. Application developers and data engineers are usually not adept at carrying out such workloads in an efficient manner. For this reason, there is a need for highly abstract reusable infrastructural components that can be easily utilized. This implies that teams should be equipped with required
945 infrastructure as a service, that can be easily employed to account for BD needs.

One way to provision such infrastructure as a service, is to utilize infrastructure as a code software tools like Terraform [124]. Besides, data infrastructure

may be extended based on currently running infrastructure for application payloads. However, this might be challenging, as the BD ecosystem is growing rapidly, and while a software application might be running in an EC2 worker node in an EKS cluster on Amazon, the BD system maybe running Databricks, or a CDP solution like Segment [125], HDFS [126], or Amazon Kinesis. This brings the challenge of composing data and application infrastructure together to provide with coherent, cost efficient interoperable infrastructure.

Nevertheless, this should not be a daunting task, as one can simply extend the worker nodes map in the EKS code written for Terraform to add a new node in the network, which installs Databricks through a Helm Chart [127]. In addition, the data infrastructure should be accompanied with proper tooling. Tools like GNU Make [128], makes it quite easy for developers and data engineers to deploy infrastructure as they demand. A mature infrastructure as a service should provide the team with core infrastructures such as BD storage, stream processing services, batch processing services, event backbones or message queues, and data integration technologies.

6.1.5. Governance through a federation service

The other principle of Cybermycelium is the global governance or the global standardization of the services. This principle is perhaps a lesson learnt from the studied application of Miroservices architecture in the industry [21]. Distributed architectures are made up of independent collection of nodes, with distinct lifecycle that are deployed separately and are owned by various teams. As the number of these services grow, and the interconnections increase, the challenge of maintaining and scaling the system increases. This means services need to interoperate, provide services, ingest data from other services, perform graph or set operations in a timely manner and do stream processing.

In order to scale and maintain these independently deployed yet interconnected services, Cybermycelium needs a governance model that embrace domain autonomy, distribution and decentralization, automation and Devops, and increased interoperability through federated government. This requires a shift

in thinking, which obsoletes many prevalent assumptions of software and data engineering. The point of federation is not to suppress or kill the creativity and innovation of the teams, but rather, introduction of global contracts and standards that are in-line with company's resources and vision. This therefore can be a challenging task for software architects to find the equilibrium between right amount of centralization and decentralization. For instance, semantic related metadata can be left to the product domain to decide, whereas policies and standards for metadata collection should be global. This is somewhat analogous to architectural principles in TOGAF's ADM [129].

The definition of these standards is up to the architecture, or architectural governance group, and is usually achieved through service level objectives (SLOs) or well-defined contracts and standards.

6.1.6. *Event driven services*

Cybermycelium has been designed in a decentralized and distributed manner. Despite the advantages of decentralized systems in terms of maintenances and scalability, communication between the services remains a challenge. As the number of services grow, the communication channels increases, and this soon turns into a nexus of interconnected services that each try to meet its own end. Each service will need to learn about the other services, their interfaces, and how the messages will be processed. This increases the coupling between services and makes maintenance a challenging task. We argue that this should not be the aim of a distributed RA such as Cybermycelium.

One approach to alleviate these issues is asynchronous communication between services through events. This is a different paradigm to a typical REST style of communication. A point-to-point communication occurs between services as series of 'command', like getting or updating a certain resources, whereas event-driven communication happens as a series of events. This implies that instead of service A commanding service B for certain computation, service B reacts to a change of state through an event, without needing to know about service A.

This provides a 'dispatch and forget' kind of a model, in which a service is only responsible to dispatch an event to a topic of interest for the desired
1010 computation. In this way, the service does not need to wait for the response and see what happens after the event is dispatched, and is only responsible for dispatching events through a well-defined contract. Underlying this paradigm, services do not need to know about each other, but rather they need to know what 'topic' they are interested in.

1015 This is analogous to a kitchen, where instead of a waiter needing to communicate directly to another waiter and to the chef and to the cook, they all react to certain events, such as customers coming in, or an order slip being left on a counter. The subtlety lies in the underlying paradigm and philosophy of 'event' instead of 'command'. This paradigm solves many issues of communication in distributed BD systems such as long running blocking tasks, throughput,
1020 maintenance, scale and ripple effect of service failure.

It is worth mentioning, that eventual consistency (BASE) is preferred over ACID transactions for performance and scalability reasons. The detail of these two varying kind of transactions is outside the scope of this study.

1025 6.2. The Artefact

After having discussed many kernel and design theories, the necessary theoretical foundation is created for the design and development of the artefact. Cybermycelium is created with Archimate and displays the RA mostly in technology layer. Displaying these services in technology layer means that it's up
1030 to the designer to decide what flow and application should exist in each node. For the sake of completion, and as every software is designed to account for a business need, we have assumed a very simple BD business process. While this business layer could vary in different context, Cybermycelium should be able to have the elasticity required to account for various business models. To ease
1035 understanding of the RA, we sub-diagrammed in product domain (6).

Cybermycelium is made up of 11 main components and 9 variable components as depicted in figure 5.

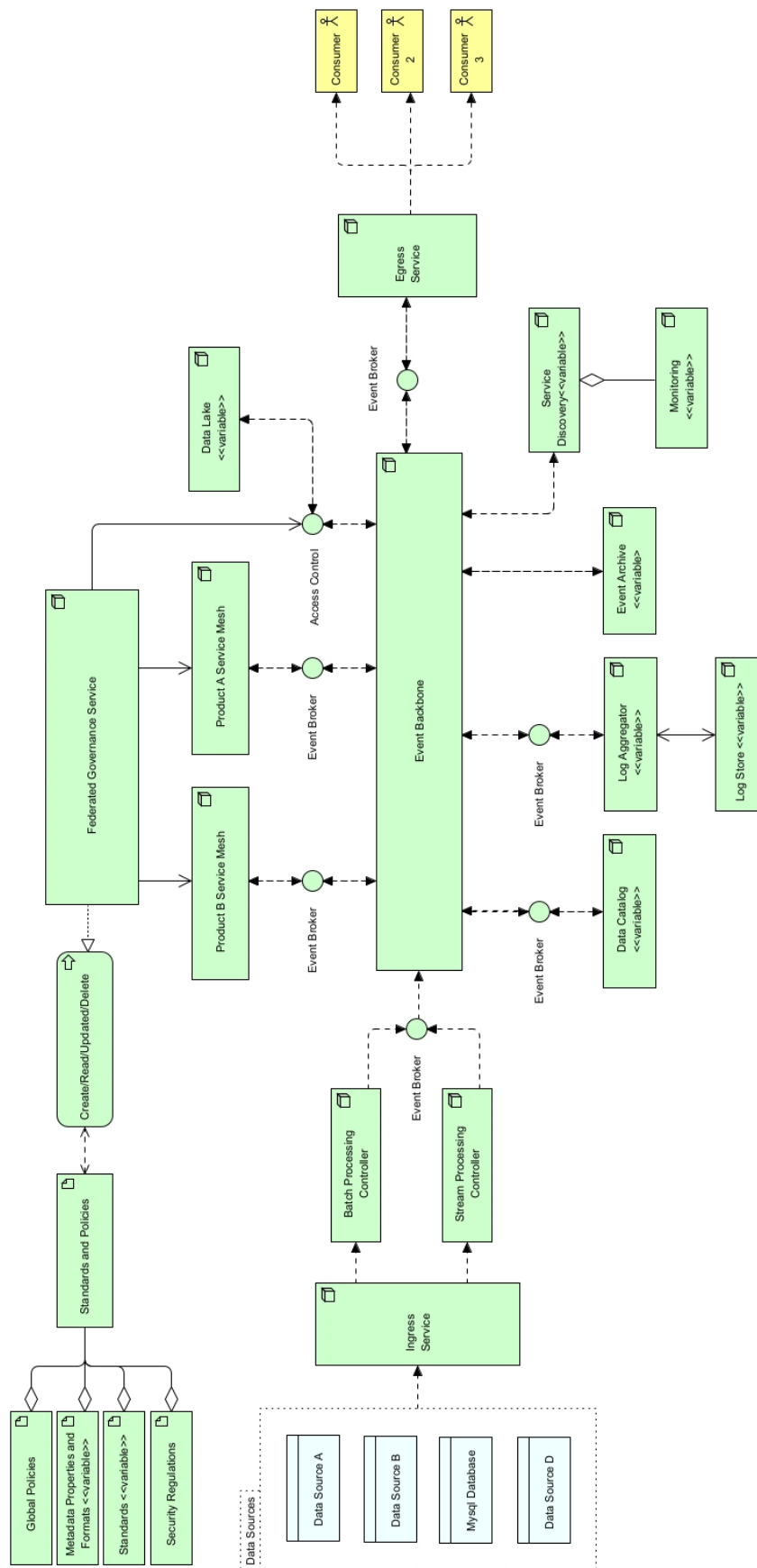


Figure 5: Cybermycelium BD Reference Architecture

For the purposes of effective modeling and abstraction, we've sub-diagrammed the product domain, which is portrayed in figure 6

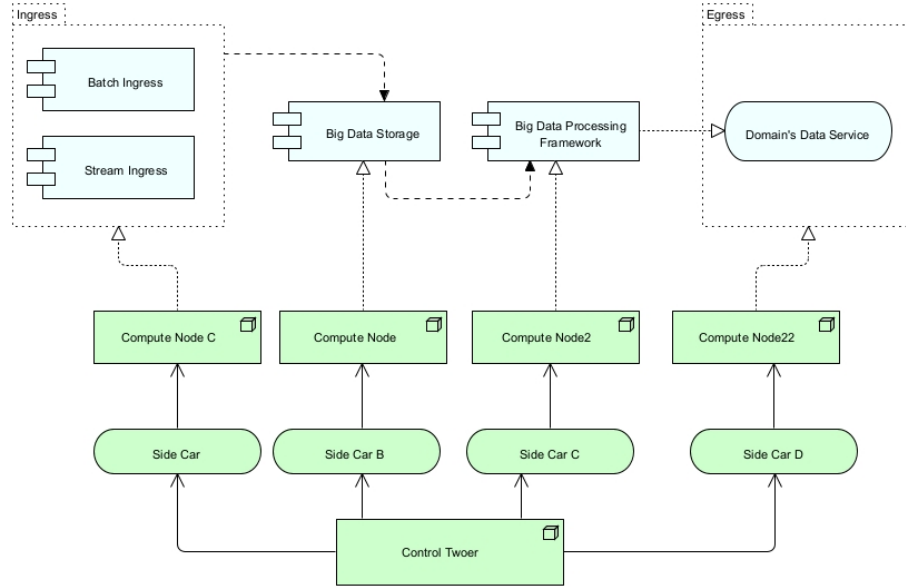


Figure 6: Cybermycelium Product Domain Design

The main elements are;

1. **Ingress Service:** The ingress service is responsible for exposing the necessary port and endpoint for the data to flow to the system. Depending on the nature of the request, the ingress service will load balance to either batch processing controller or stream processing controller. It is essential for the ingress service to operate asynchronously to avoid any potential choke points. In addition, ingress handles the SSL termination, and potentially name-based virtual hosting. Ingress has several benefits. Firstly, it helps with security by preventing port proliferation and direct access to services. Secondly, it help with performance by distributing requests based on their nature, and SSL termination. Thirdly, if there's a need for object mutation through a proxy, ingress is the best architectural construct. Having an ingress also means that the point of entry is clear, which makes monitoring easier, and allows for other components of the architecture to

remain in private networks. This component addresses the requirements
1055 Vol-1, Vol-2, Var-1, Var-3, Var-4, Val-1, Val-3, Val-4, SaP-1 and SaP-2.

2. **Batch Processing Controller:** Batch processing controller is responsible for dispatching batch events to the event backbone. This service should be a small service (could be a Lambda) with the main responsibility of receiving a request for batch processing and dispatching an event to the
1060 event broker. Because the nature of the request is of type batch and has been clearly distinguished by the ingress, batch processing controller can dispatch events in bulk and asynchronously. This is the main difference of this service to stream processing controller. Batch processing controller can execute other non compute-intensive tasks such as scrubbing properties from the given data or adding headers. Having a specific controller
1065 for batch processing improves monitoring and allows for customized batch event producing. This component addresses the requirements Vel-1, Val-1, and Val-2.

3. **Stream Processing Controller:** Stream processing controller is responsible for dispatching streaming events to the event backbone through the
1070 event broker. This service has been segregated from the batch service as it has to account for a different nature of events. Streams are synchronous in nature, and can require high-throughput. This service is a small service as well, but non-heavy computations such as enabling stream provenance, and one-pass algorithms can be utilized. Having a specific controller for
1075 streaming processing means that custom attributes can be associated to stream events, and the events can potentially be treated differently based on the nature of the system. This also eases monitoring and discovery. This component addresses the requirements Vol-1, Vel-1, Vel-2, Vel-4, Vel-
1080 5, Val-2,

4. **Event Broker:** Event brokers are designed to achieve 'inversion of control'. As the company evolves and requirements emerge, the number of nodes or services increases, new regions of operations may be added, and new events might need to be dispatched. As each service has to com-

1085 municate with the rest through the event backbone, each service will be
 required to implement it's own event handling module. This can easily
 turn into a spaghetti of incompatible implementations by various teams,
 and can even cause bugs and unexpected behaviors. To overcome this
 challenge, an event broker is introduced to each service of the architec-
 1090 ture. Each service connects to its local event broker and publishes and
 subscribes to events through that broker. One of the key success criteria
 of the event broker is a unified interface that sits at a right level of abstrac-
 tion to account for all services of the architecture. Event brokers, being
 environmentally agnostic can be deployed to any on-premise, private or
 1095 public infrastructure. This frees up engineers from having to think about
 the event interface they have to implement and how it should behave.
 Event brokers can also account for more dynamism by learning which
 events should be routed to which consumer applications. Moreover, event
 brokers do also implement circuit breaking, which means if the service
 1100 they have to broke to is not available and does not respond for a certain
 amount of time, the broker establishes unavailability of the service to the
 rest of the services, so no further requests come through. This is essential
 to preventing a ripple effect over the whole system if one system fails. This
 component indirectly addresses the requirements Val-1, and Ver-1.

1105 5. **Event Backbone:** This is the heart of the Cybermycelium, facilitating
 communication among all the nodes. Event backbone in itself should be
 distributed and ideally clustered to account for the ever-increasing scale of
 the system. Communication occurs as choreographed events from services
 analogous to a dance troupe. In a dance troupe, the members respond to
 1110 the rhythm of the music by moving according to their specific roles. In
 here, each service (dancer) listens and reacts to the event backbone (music)
 and takes the required action. This means services are only responsible
 for dispatching events in a 'dispatch and forget' model, and subscribe to
 the topics that are necessary to achieve their ends. Event backbone thus
 1115 ensures a continues flow of data among services so that all systems are

in the correct state at all times. Event backbone can be used to mixed several stream of events, caching of events, archiving of events, and other manipulation of events, so long as it's not too smart! or does not become a ESB of SOA architectures. Ideally, an architect should perceive the event backbone as series of coherent nodes that aim to handle various topics of interest. Over the time, event backbone can be monitored for access patterns and facilitate the communication at a faster rate. This component addresses the requirements Vel-1, Vel-2, Vel-3, Vel-4, Vel-5, Val-1, Val-2, Ver-1, Ver-2, and Ver-3.

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6. **Egress Service:** The egress service is responsible for providing necessary APIs for the consumers of the system to request data in demand. This is a self-serve data model in which data scientists or business analyst can readily request data from various domains based on the data catalogue. Clients can first request for a data catalogue and then use the catalogue to request for the product domain that accounts for the desired data. This request can include several data products. Egress is responsible to route the request to the data catalogue, and to corresponding product service mesh' in order to resolve values. The egress realizes the address to service meshes and other services through the data catalog and service discovery.

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The egress service should cache the resolved addresses and values in order to increase performance and response time. An architect can even choose to implement a complete query cache component inside the egress service, however that will make maintenance potentially more difficult and may result in bottlenecks. This is to avoid having people requesting directly to data engineers for various big data requirements, and means that people can just request for what data they need, analogous to a person who orders food at a restaurant, menu being the data catalog, and egress being the waiter. This component addresses the requirements Vel-2, Vel-4, Val-3, Val-4, SaP-1, and SaP-2.

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7. **Product Domain Service Mesh:** As previously discussed a product is a capability of the business, and each product has its own domain consisting

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of the bounded context and the ubiquitous language. In system and architectural terms, these domains are implemented as a service mesh. Each service mesh is made up of a batch ingress, stream ingress, big data storage, big data processing framework, domain's data service, the required compute nodes to run these services, a side car per service and a control tower. These components provide the necessary means for the domain to achieve its ends. This architectural component removes the coupling between the teams and promotes team autonomy. This means people across various teams are enhanced with the desired computational nodes and tools necessary, and can operate with autonomy and scale without having to be negatively affected by other teams or having friction with platform teams or siloed data engineering teams. This component indirectly addresses Vol-1, Vel-3, Vel-4, Vel-5, Var-1, Var-2, Var-3, Val-1, Val-2, Val-3, Val-4, Sap-1, SaP-2, Ver-1, Ver-2, and Ver-3.

8. **Federated Governance Service:** Evidently, Cybermycelium is a distributed architecture that encompasses variety of independent services with independent lifecycle that are built and deployed by independent teams. Whereas teams have their autonomy established, in order to avoid haphazard, out-of-control and conflicting relations, there should be a global federated governance that aims to standardize these services. This will facilitate the interoperability between services, communication, aggregates, and even allows for a smoother exchange of members across teams. This also means the most experienced people at a company such as technical leads and lead architects will prevent potential pitfalls that more novice engineers may fall into. However the aim of this service is not centralize control in anyway, as that would be going a step backward into the data warehouse era. The aim of this service is to allow autonomous flow in the river of standards and policies that tend to protect company from external harm. For instance, failing to comply to GDPR while operating in europe can sets forth fines up to 10 million euros, and this may not be something that novice data engineers or application developers are fully

aware of. The real challenge of the governance team is then to figure out the necessary abstraction of the standards to the governance layer and the level of autonomy given to the teams. The federated governance service is made up of various components such as global policies, metadata elements and formats, standards and security regulations. These components are briefly discussed below;

- (a) **Global Policies::** general policy that govern's the organizational practice. This could be influenced by internal and external factors. For instance, complying to GDPR could be a company's policy and should be governed through the federated governance service.
- (b) **Metadata Properties and Formats::** this is an overarching metadata standard defining the required elements that should be captured as metadata by any service within the organization, the shape of and the properties surrounding it. For instance, the governance team may decide that each geographic metadata should conform to ISO 19115-1 [130].
- (c) **Standards:** overall standards for APIs (for instance Open API), versioning (for instance SemVer), interpolation, documentation (for instance Swagger), data formats, languages supported, tools supported, technologies that are accepted and others.
- (d) **Security Regulations:** company wide regulations on what's considered secured, what softwares are allowed, how interfaces should be conducted and how the data should be secured. For instance, company may choose to alleviate risks associated with OWASP top 10 application security risks.

This component can indirectly affect all requirements.

- 9. **Data Catalog:** As the products increases , more data become available to be served to consumers, interoperability increases, and maintenance becomes more challenging. If then, there is no automatic way for various teams to have access to the data they desire, a rather coupled and slow BD culture will evolve. To avoid these challenges and to in crease discoverabil-

ity, collaboration, and guided navigation, the service data catalog should be implemented. Data catalog has been listed as a must-have by Gartner [131] and introduces better communication dynamics, easier data serve by services and intelligent collaboration between services. This component addresses the requirements Vel-4, Var-1, Var-3, and Var-4.

10. **Logging Aggregator and Log Store:** If all services employ the idea of localized logging, and simply generate and store logs in their own respective environments, debugging, issue finding and maintenance can become a challenging task. This is due to the distributed nature of Cybermycelium and the requirements to trace transactions among several services. In order to overcome this challenge, we've employed the log aggregator pattern popularized by Chris Richardson [132]. The log aggregator service is responsible for retrieving logging events through the event broker from individual services and write the collected data into the log store. The log aggregator configuration and semantics is up to the designer and architecture team. This allows for a distributed tracing, and graceful scaling of organizational logging strategies. This component indirectly addresses the requirements Vol-1, Vel-1, Val-1, and Ver-1.
11. **Event Archive:** As the quantity of services grow, the topics in event backbone increases, and the number of events surges. Along the lines of these events, there could be a failure, resulting in timeout and a loss of series of events. This brings system in a wrong state and can have detrimental ripple effect on all services. Cybermycelium tends to handle these failures by using an event archive. The event archive as the name states, is responsible for registering events, so they can be retrieved in the time of failure. If there was a blackout in certain geographical location and the event backbone went down, the backbone can recover itself and bring back the right state of the system by reading the events from the event archive. Th event broker is responsible for circuit breaking, so the service do not request any more events to the backbone while its down. The time to expiry, and what events should be archived is decided based

1240 on context in which Cybermycelium is implemented. This component indirectly addresses the requirements Vol-1, Vel-1, Val-1, and Ver-1.

12. **Data Lake:** Whereas Cybermycelium is a great advocate of decentralized and distributed systems, we do not find it necessary for each product domain to have it's own kind of a data lake or data storage. This is to prevent duplication, contrasting data storage approaches, decreased oper-
 1245 ability among services and lack of unified raw data storage mechanisms. Data lake has been designed to store large volume of data in raw format before it can get accessed for analytics and other purposes. This means data can be first stored in the data lake with corresponding domain own-
 1250 ership before it needs to be accessed and consumed by various services. Structured, semi-structured, unstructured and psudo-structured data can be stored in data lake before it gets retrieved for batch and stream processing. Nevertheless, this does not imply that all data should directly go to the data lake; the flow of data is determined based on the particularities
 1255 of the context in which the system is embodied. One approach that we find suitable is for each team to own a unit of storage in the data lake, which is handled by the access control as depicted on the model. This component addresses the requirements Vol-2, Vel-1, Var-1, Var-3, Var-4, Val-3.

1260 13. **Service Discovery:** In a distributed setup like Cybermycelium, how do services discover the location of other services? This is achieved through service discovery. As the practice of hard-coding service addresses in con-
 1265 figuration files is not a maintainable or scalable approach, one has to think about an automated scalable solution in which services can become discoverable by other services. The service discovery node is responsible for this job. This is achieved through services registering themselves to the service discovery node when they boot up. Service discovery then ensures that it keeps an accurate list of services in the system, and provides the API necessary for others to learn about the services. For instance, it's
 1270 idiomatic for an engineer to specify a command to be executed when a

Docker container starts (*Node server.js*); thus one can imagine extending the boot up instructions to achieve the registration to the service discovery node. This somewhat resembles to DHCPs and house wifi networks. This component indirectly addresses the requirements Vel-2, Vel-4, Var-2, Var-4, Val-3, Val-4, SaP-2.

14. **Monitoring:** Monitoring systems are integral to robustness of highly dynamic ecosystem of distributed systems and directly affect metrics such as mean time to resolution (MTTR). Services emit colossal amounts of multi dimensional telemetry data that covers a vast spectrum of platform and operating system metrics. Having these telemetry data captured, handled and visualized, helps systems engineers, software reliability engineers, and architects proactively address upcoming issues. Based on these premises, the main responsibility of this service is to capture and provide telemetry data from other service to increase the overall awareness of the Cybermycelium ecosystem. This service is tightly aggregated with the service discovery. Monitoring services help storing these data to fuel proactive actions. This component indirectly addresses all requirements.

The variable elements in Cybermycelium can be adjusted, modified and even omitted based on the architect's decision and the particularities of the context. The aim of this RA is not limit the creativity of data architect, but to facilitate their decision making process, through introduction of well-known patterns and best practices from different school of thoughts. While we still recommend keeping the variable components, an architect may decide to embark on a more complicated metadata approach rather than just a data catalog. We do not elaborate on all alternative options for each variable module as industry constantly changes, and architects constantly aim to design systems that address the emerging problem domains.

7. Evaluation:

Of particular importance to development of a RA, is the evaluation of it.
1300 As discussed earlier, we aim to evaluate the correctness and utility of the RA
by how it can be turned into an effective context-specific concrete architecture,
following the guidelines of ATAM. The main goal of ATAM is to appraise the
architectural decisions and their consequences in light of quality attributes. This
method ensures that the architecture is under the right trajectory and in-line
1305 with the context. Architecture is a an amalgamation of risks, tradeoffs, and
sensitivity points. Using ATAM increased our confidence by uncovering key
architectural tradeoffs, risks, and sensitivity points.

For ATAM to be successful there should not be a precise mathematical
analysis of system’s quality attributes, but rather trends should be identified
1310 where architectural patterns are correlated with a quality attribute of interest.
For brevity purposes, we do not expand on what ATAM is, and the details of
each steps in it, and we only explain how the evaluation has been conducted
through ATAM. It is important to note that, this wasn’t a setup in which an
outside evaluation team would come to a company to evaluate an architecture
1315 in practice, but it was our prototype that we brought into a company to test
its utility and relevance. While we could have achieved this with technical
action research or leightweight architecture evaluation, we found ATAM to be in-
line with our conceptual constructs, which are architectural constructs. ATAM
provided us with a framework to discuss architectural concepts in a rigorous
1320 way [133]. Howbeit, to avoid bias, we invited a third-party researcher who is
familiar with ATAM to observer the overall process.

For instantiation of the RA, We utilized ISO/IEC 25000 SQuaRE standard
(Software Product Quality Requirements and Evaluation) [134] for technology
selection. We did not fully adopt this standard, but was rather inspired by it to
1325 make a better decision. The quality model in the standard is based on charac-
teristics, sub-characteristics and standards. This standard also references many
other standards for maintenance and quality keeping of computer systems. The

detailed explanation of the standards and its constituent elements are outside the scope of this study.

1330 By applying standard and the requirements of Cybermycelium to the pool of available tools in the industry, we successfully created the archetype. We chose the tools are the mostly adopted and do support the architectural requirements of Cybermycelium. We did not want to develop tools from scratch, as that would delay our evaluation artefact and this would affect the stakeholders negatively. In addition, many mature tools exist that satisfy our architectural requirements, so therefore 'reinventing the wheel' was unnecessary.

We chose Node JS for all APIs and custom scripting, Nginx as our ingress, AWS Lambdas for stream and batch processing controllers, Kafka for event backbone, Kafka event brokers as the event broker, AWS application load balancer as the egress load balancer, Istio as the control tower, Envoy as the side car, Kubernetes as the container orchestrator, AWS S3 as the BD store and event archive, and Data Bricks for stream and batch processing. The concrete architecture created out of Cybermycelium for the archetype is depicted in figure 7

1345 We aimed to incorporate most components of our RA into this instance, however logging, monitoring, service discovery, federated governance service, and data catalog has been omitted. Some details of this evaluation is omitted to protect the security, and intellectual property of the practice, and some details are modified for academic purposes. These modifications have not affected the integrity of the evaluation. This evaluation is an iterative process, collecting feedback from different group of stakeholders in each phase.

7.1. Phase 1:

This evaluation is undertaken in a subsidiary of a large-scale international company that has over 6000 employees all around the globe. The subsidiary company offers a practice management software for veterinary practitioners via Software as a Service (Saas) and has over 15,000 customers from the USA, UK, Australia, New Zealand, Canada, Singapore and Ireland, among which

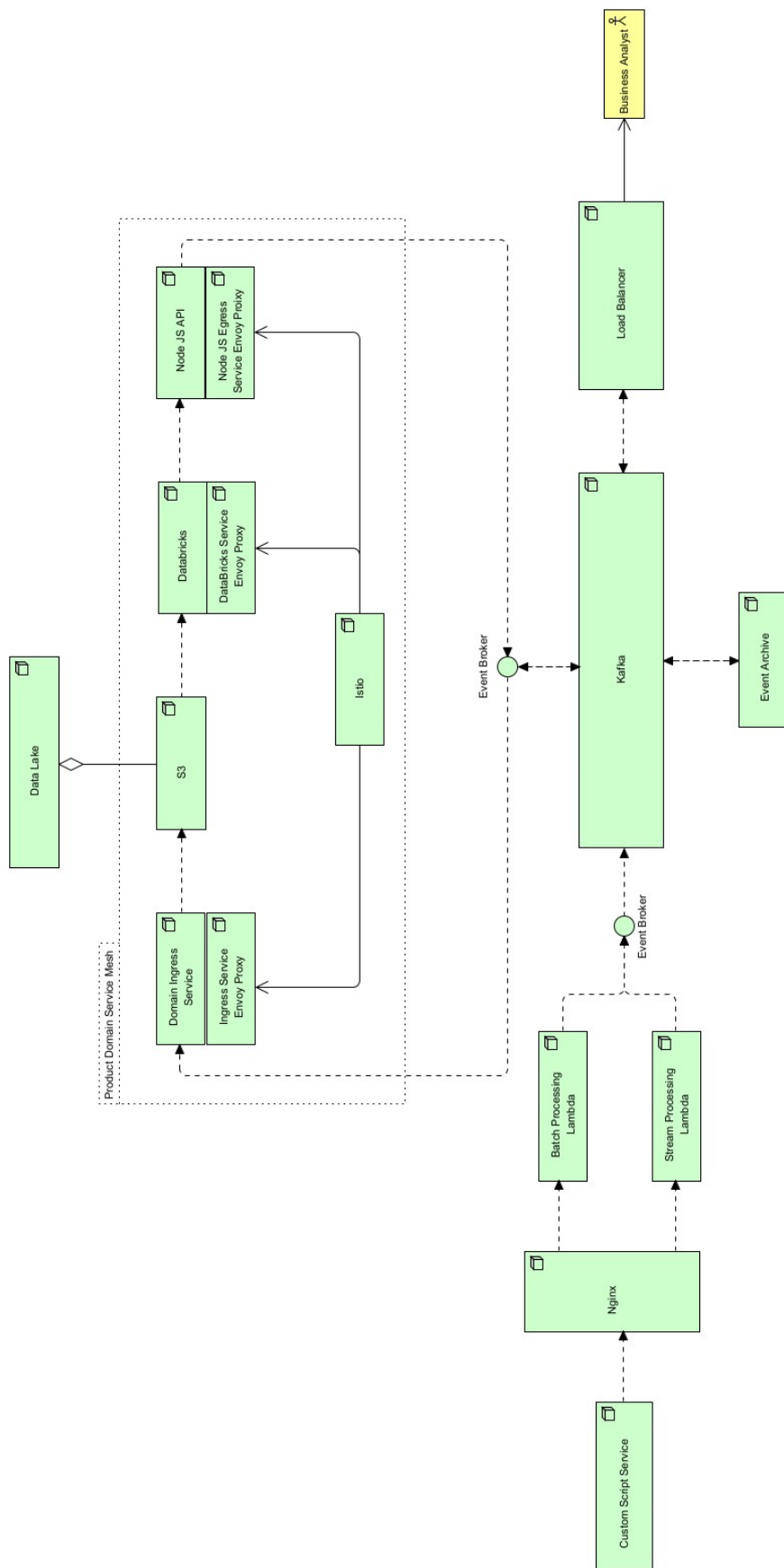


Figure 7: Cybermycelium Instantiation

are some of the biggest equine hospitals, universities and veterinary practices. The company is currently at the stage of shifting from centralized synchronous
1360 architecture into decentralised event driven microservices architecture, and is ambitions to adopt artificial intelligence and BD.

The initial step was the identification of relevant stakeholders. For this purpose we have approached the key stakeholders in the company's technical governance team. Our aim was to incorporate at least two lead architects of
1365 the company in this process. We emphasized on architects that have been with business for a long period of time. This was to ensure that we do not miss any important element in the process of evaluation. As a result, we invited two lead development architects, head of product, and a business analyst for phase 1.

7.1.1. Step 1 and 2: Introduction

During the initial meeting, in step 1, ATAM was presented with clear description of its purposes. In step 2, stakeholders discussed the background of the business and some of the challenges faced, and the current state of affairs, the primary business goals, and architecturally significant requirements. This step illuminated on integral elements such as: 1) most important functions of
1375 the system, 2) any political, regional, or managerial constraints, 3) the business context and how it relates to our prototype, 4) architectural drivers.

7.1.2. Step 3: Present the Architecture

In step 3, the prototype has been presented, our assumptions have been stated, and variability points portrayed.

7.1.3. Step 4: Identifying Architectural Approaches

To establish the architectural styles, we first analyzed the archetypes with respect to key quality attributes chosen for this evaluation which are performance, availability and modifiability.

- For performance, Nginx, Kafka, Istio, DataBricks and the AWS Application Load Balancer have been described.
-
- 1385

- For availability, Kafka, Event archive, Nginx, controllers, Data Lake and Istio have been discussed.
- For modifiability, the concept of domain driven design, the service mesh, zero coupling, the plug and play nature of the archetype, the ability to add desirable services through event brokers, and the distributed nature of the architecture has been discussed.

We then probed each of the quality attributes for potential risks, tradeoffs, and sensitivity/variability points.

7.1.4. Step 5: Utility Tree Elicitation:

In order to generate the utility tree, first we had to learn what are the most important quality attributes. While we learnt about these quality attributes in step 2, in this step we probed deeper. We first presented our assumptions, and double-checked it with the stakeholders. Whereas some stakeholders raised concerns over privacy, the members unanimously agreed that performance, availability, and maintainability are the most important quality attributes. This was in-line with our assumptions. In this process, we rated the technical difficulty, and the key stakeholders rated the business importance.

Based on these premises, the utility tree has been generated (figure 8).

7.1.5. Step 6: Analyze Architectural Approaches:

After having scenarios prioritised and architectural approaches identified, it's time to analyze if the architectural approaches are the fitting for the given scenarios. In this step, we asked the lead architects to probe the architecture and we explained how the prototype is addressing each scenario.

The result of this step, was identification of sensitivity points, trade-offs, risks and non-risks. This step took longer than what we anticipated, as variety of questions arose and many aspects of the architecture was challenged. The detail is discussed in 'presenting the results' step.

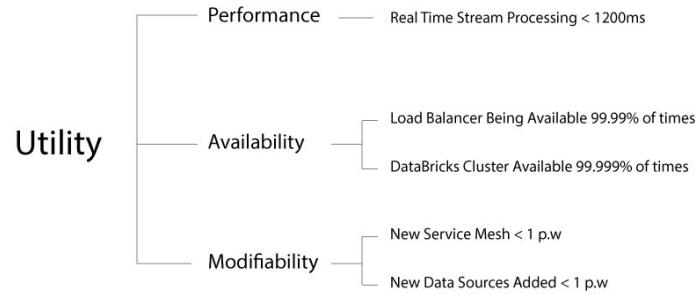


Figure 8: Utility Tree

7.2. Phase 2:

This phase was a more serious phase of our evaluation, as we invited more stakeholders, collected more scenarios and even created simulations. For this phase, in addition to lead architectes, we invited a product owner responsible for the product in which the artifact is tested, a quality assurance engineer and several developers. We repeated step 1, and provided with recap of step 2 to 6, and shared the current list of risks, non-risks, sensitivity points, and tradeoffs.

This phase is an iteration of phase one, so we collected scenarios, analyzed architectural approaches, and finally presented the result of the evaluation.

7.2.1. Step 7: Brainstorm and Prioritize Scenarios

While utility tree was created in step 5, the purpose of this brainstorming was to get a pulse of the larger stakeholder community and capture more scenarios.

Scenarios are the quantum of ATAM, and help captring stimuli to which the architecture has to react. These stimuli help to highlight system’s ability to too meet desired functional and non-functionanl requirements [107]. Based on this premise, in this step, we asked stakeholders to come up with three different kind of scenarios namely growth scenarios (anticipated changes), use-case scenarios

1430 (typical usage of the system) and exploratory scenarios (extreme cases). The
result of this created 20 scenarios, which we then asked stakeholders to vote
on. The result of the voting yielded 5 scenarios which are described as two user
journeys;

- A cat owner brings a cat to the veterinary hospital. The cat has symptoms
1435 of a lyme disease, and should be diagnosed in a timely manner to avoid
master problems.
- There has been numerous cases of cancer in pets in certain environments.
This environment should be analyzed to see if environmental factors play
a cancer inducing role.

1440 7.2.2. Step 8: Analyze Architectural Approaches

Before starting this step, we took a few days break to simulate the scenarios
against our prototype. While ATAM does not prescribe this, we augmented
our evaluation with this simulation to ensure that we do not overlook any nec-
essary architectural detail. This improved our confidence in our RA and the
1445 architectural probing questions to come.

We emulated the scenarios against the prototype by creating relevant topics
in the Kafka, having the data flow, having the ingress in the service mesh digest
it and flow it into the storage and processing and so on so forth. We've been us-
ing real world data, so there was no need for data fabrication and synthesis. We
1450 configured Nginx to pass the request to the responsible Lambdas, and Lambdas
then produced the necessary events and sent it to Kafka.

We presented our simulation alongside some metrics we captured and dis-
played in our cloud served Grafana instance. From here on, this step followed
the exact same procedure as step 6, with a difference that this time there's
1455 been more extensive probing and analysis of the architecture and the simulated
scenarios. The simulation and results of it helped clarifying on some of our
architectural constructs and led to emergence of several questions:

- How does the system recover if the event backbone goes out of order?

- What if the service mesh ingress is not available?
- 1460 • Should privacy be it's own service? or should it sit in federation ?
- Should have a dedicated service mesh for metadata management?
- How easy it is to extend and modify current services?
- Should there be a certain order to events ?
- Is there a benefit in creating event mesh between event brokers?
- 1465 • Where is the best place to scrub sensitive data from the incoming streams?

7.2.3. Step 9: Present Results

In this last step, the collected theories from the process of evaluation is discussed in terms of quality attributes, risks, sensitivity points, tradeoffs and other unplanned discussions that arose during the meetings.

1470 Based on the result of our evaluation, stakeholders feedback, utility tree, and the architectural qualities of Cybermycelium, we deduce that system quality Q_S , is a function f of the quality attributes availability Q_A , performance Q_P , and modifiability Q_M .

$$Q_S = f(Q_M, Q_A, Q_P) \quad (1)$$

Performance:. In order to analyze our approach in-line with the utility tree, 1475 after having created the simulated scenarios, we used a cloud stress testing testing agent (StressStimulus). After having run this stress test a couple of times, it has become evident to us that cold start latency of AWS Lambda services can affect the performance requirements stated in the utility tree. A Lambda can take anywhere from 100ms to over a second on cold start time. 1480 This latency varies and hard to nail down, but even considering the latency, we have captured an average of 1000ms response time from our system which is inline with the utility tree. While replacing Lambdas with EC2s or Fargates could solve this issue, it would increase the cost, affect the maintainability of the

architecture (a server has to be provisioned and maintained), and would require
1485 rework of several services.

8. Discussion

Our findings from this study yielded the fact that progress is uneven in the area of BD RAs. While there are many researches in the area of data warehousing, artificial intelligence, data science, and IOT, data engineering seems to
1490 be needing more research. While, there are many well established approaches for crunching large volume of data, or handling dimensionality of complex data sets, the overall organization of BD technologies, which is the architecture, needs more attention from academia and industry.

Majority of the BD RAs that we have analyzed were running underlying some
1495 sort of a monolithic data pipeline with a central storage. This is a challenging architecture to scale and maintain. How does one takes preventive measures to stop a data lake from turning into a data swamp ? How a team of hyper-specialized siloed data engineers that are running the data pipelines, will be aware of the actual business problem and therefore keep a certain level of quality
1500 of that data? how data interoperability is achieved? how data ownership is institutionalized ?

If a software engineers decides to, for instance, manipulate a certain field in a certain entity's schema for the development of a new feature, how will this affect the data engineering process and how is this communicated? as data becomes
1505 more and more available to the company, the ability to consume it all in one place diminishes.

On the other hand, the current state of BD RAs do not seem to be very far away from traditional data warehousing approaches. In fact, some of them have adopted the idea of data marts and propose them as BD solution, but using
1510 newer technologies. Moreover, some architectures have attempted to utilize data lake to serve data analysts and business intelligence.

We posit that neither the attempt to onboard BD analytics workloads to

data warehouses, nor the attempt to serve business intelligence with data lakes is gonna result in a scalable and maintainable system. We therefore propose the
1515 need for future research directions in the area of decentralized and distributed BD RAs.

We also felt that the quality of many of BD RAs published does not seems to be enough to meet the industrial expectations. This is due to the challenges of developing BD RAs and the cost and resources required to evaluate these
1520 artifacts. It is also worth mentioning that a rigorous methodology for evaluating reference architectures are quite rare, and while there are studies that have attempted to address these issues ([50]), there is a need for more research in this area.

Given all, we posit that RAs can be considered an effective initial point to design and development of BD systems. These artifacts helps facilitating communication, capture requirements from various stakeholders, and catch design
1525 issues while they are still cheap. Based on this, therefore, more and more attention needs to be given to this area and its foundational methodological needs. This study does not aim to do a deep comparison of Cybermycelium and other
1530 RAs, the major architectural constructs, the challenges associated to current BD RAs, and the reasoning behind our artifact should elucidate the varying nature of our artifact.

9. Comments to address

1. The explanation of the ATAM in §6 is not correct. Phase 1 and Phase
1535 2 are iterations, using different stakeholders and techniques to elicit scenarios, but performing the same analysis. You might have picked up this misinterpretation from Ford and Richards, which indicates that maybe you are referencing work (i.e., Ref. 32) that you did not actually read.
2. The authors validate the proposed reference architecture by creating a
1540 prototype and using ATAM to evaluate them. It would be interesting to know how the prototype has been designed, how the RA helped in the

definition of the prototype, etc.

3. In addition, in the validation, the authors propose different metrics for the performance, availability, and modifiability, but the results for the implemented prototype are not provided.

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10. Conclusion

Data engineering is a complicated endeavour, and while there are many good practices for service distribution in software engineering, the BD domain does not seem to benefit from all of these ideas. This has made BD system development a daunting task, and many companies have failed to bring to light the potential of data-driven decision making. Therefore, there is more and more research required in the areas of data architecture, and the ways in which the data flows between various components. RAs are a good start to such complicated tasks. By absorbing the best of knowledge from the practice and injecting it as a living model into practice, practitioners can benefit from already identified pitfalls. BD systems have got a long way to mature, but with clear direction both in the industry and academia, this aspiration can come to fruition in near future.

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Appendix A. Studies included in the systematic literature review

Study	Author	Year	Type
Towards a big Data reference architecture	[103]	2013	Master’s Dissertation
A reference architecture for Big Data solutions introducing a model to perform predictive analytics using Big Data technology	[135]	2013	Conference Paper

IBM - Reference architecture for high performance analytics in healthcare and life science	[136]	2013	Book
Big data ecosystem reference architecture	[137]	2013	White Paper
A proposal for a reference architecture for long-term archiving, preservation, and retrieval of Big Data	[138]	2014	Conference Paper
Questioning the Lambda architecture; Kappa Architecture	[59]	2014	Blog
Defining architecture components of the Big Data Ecosystem	[139]	2014	Conference Paper
Oracle - Information Management and Big Data: A Reference Architecture	[140]	2014	White Paper
Big Data driven e-commerce architecture	[141]	2015	Journal Article
The solid architecture for real-time management of big semantic data	[142]	2015	Journal Article
Reference architecture and classification of technologies, products and services for big data systems	[143]	2015	Journal Article
A Reference Architecture for Big Data Systems	[144]	2016	Conference Paper

A reference architecture for Big Data systems in the national security domain	[145]	2016	Conference Paper
A Reference Architecture for Supporting Secure Big Data Analytics over Cloud-Enabled Relational Databases	[146]	2016	Conference Paper
SAP - NEC Reference Architecture for SAP HANA & Hadoop	[147]	2016	White Paper
Managing Cloud-Based Big Data Platforms: A Reference Architecture and Cost Perspective	[148]	2017	Journal Article
Scalable data store and analytic platform for real-time monitoring of data-intensive scientific infrastructure	[148]	2017	PhD Dissertation
A software reference architecture for semantic-aware Big Data systems; Bolster Architecture	[30]	2017	Journal Article
Simplifying big data analytics systems with a reference architecture	[75]	2017	Conference Paper
ISO/IEC/IEEE 42010:2011	[82]	2017	Standard
NIST Big Data Interoperability Framework: Volume 6, Big Data Reference Architecture	[86]	2018	White Paper

Towards a secure, distributed, and reliable cloud-based reference architecture for Big Data in smart	[24]	2019	Book Chapter
Reference Architectures and Standards for the Internet of Things and Big Data in Smart Manufacturing	[149]	2019	Conference Paper
Lambda architecture	[150]	2019	Conference Paper
ISO/IEC 20546:2019	[76]	2019	Standard
ISO/IEC TR 20547-1:2020	[77]	2020	Standard
NeoMycelia: A software reference architecture for big data systems	[3]	2021	Conference Paper
Smart transportation: A reference architecture for big data analytics	[74]	2021	Journal Article

Table A.2: RAs and Standards found by the result of the SLR